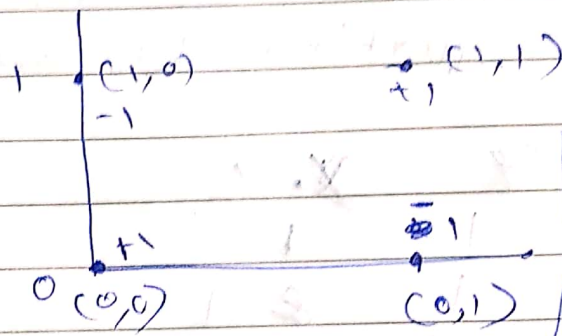


Q3)

XOR

| A | B | Output |
|---|---|--------|
| 1 | 1 | 1 |
| 0 | 1 | 0 |
| 1 | 0 | 0 |
| 0 | 0 | 1 |

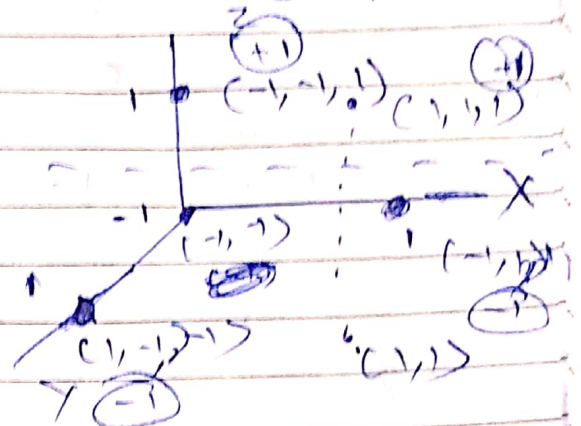


| A | B | Output |
|----|----|--------|
| 1 | 1 | 1 |
| -1 | 1 | -1 |
| 1 | -1 | -1 |
| -1 | -1 | 1 |

Assuming $0 = -1$ class -
 clearly are to figure classes are not
 linearly separable -
 \Rightarrow A kernel is needed -

3) with a new attribute $z = x \cdot y$ the equation becomes linearly separable.

| x | y | $x \cdot y$ |
|-----|-----|-------------|
| 1 | 1 | 1 |
| -1 | 1 | -1 |
| 1 | -1 | -1 |
| -1 | -1 | 1 |



$$\Rightarrow \phi \text{ Map: } (x, y) \rightarrow (x, y, xy)$$

$$\Rightarrow \text{kernel } k(x_i, x_j) = [x_i \ y_i \ x_i y_i]$$

$$\begin{bmatrix} x_j \\ y_j \\ x_j y_j \end{bmatrix}$$

$$= x_i x_j + y_i y_j + x_i y_i x_j y_j$$

Gram Matrix

$$K = \begin{bmatrix} (1,1) & (1,-1) & (-1,1) & (-1,-1) \\ (1,1) & 3 & -1 & -1 \\ (1,-1) & -1 & 3 & -1 \\ (-1,1) & -1 & -1 & 3 \\ (-1,-1) & -1 & -1 & -1 & 3 \end{bmatrix}$$

$$1 = (1, 1)$$

$$2 = (-1, 1)$$

$$3 = (1, -1)$$

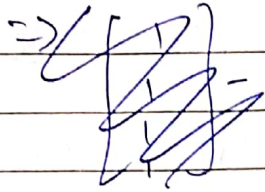
$$4 = (-1, -1)$$

$$L(x) = \sum_{i=1}^n d_i - \frac{1}{\sum_{i=1}^n d_i} \sum_{i,j=1}^n d_i d_j y_i y_j k(x_i, x_j)$$

$$= d_1 + d_2 + d_3 + d_4 - \frac{1}{2} (3d_1^2 + d_1 d_2 + d_1 d_3 - d_1 d_4 + 3d_2^2 + d_2 d_1 + d_2 d_3 - d_2 d_4 + d_3 d_1 + d_3 d_2 - d_3 d_4 + 3d_4^2 + d_4 d_1 + d_4 d_2 - d_4 d_3)$$

$$\Rightarrow \alpha_1 (3x_1 + x_2 + x_3 - x_4) =$$

$$\Rightarrow \text{in diff } \text{Lose}(x) = 1 - \frac{1}{2} H(x^T \alpha)$$



where $H = \text{Growth matrix}$
 $x = (x_1, x_2)$

$$\Rightarrow 1 - H\alpha = 0$$

$$\Rightarrow H\alpha = 1$$

$$\Rightarrow 3\alpha_1 + \alpha_2 + \alpha_3 - \alpha_4 = 1$$

$$\alpha_1 + 3\alpha_2 + \alpha_3 + \alpha_4 = 1$$

$$\alpha_1 + \alpha_2 + 3\alpha_3 + \alpha_4 = 1$$

$$-\alpha_1 + \alpha_2 + \alpha_3 + 3\alpha_4 = 1$$

$$\Rightarrow \text{optimal } \alpha_1 = \alpha_2 = \alpha_3 = \alpha_4 = \frac{1}{4}$$

\Rightarrow All are support vectors,

Sub values in Loss equation used

$$\text{Lose}(\alpha) = 1 - \frac{1}{2} \left(\frac{3}{16} x_4 + \frac{1}{16} x_4 \right)$$

$$= 1 - \frac{1}{2} \left(4 \times \frac{1}{16} \right)$$

$$= 1 - \frac{1}{2} = \frac{1}{2}$$

$$\Rightarrow \vec{w} = \sum_{i=1}^4 \alpha_i y_i \phi(x_i)$$

$$= \frac{1}{4} \left((1, 1, 1) - (-1, 1, 1) + (1, -1, -1) + (-1, -1, 1) \right)$$

$$= \frac{1}{4} (0, 0, 4) = (0, 0, 1)$$

$$\Rightarrow \text{boundary} = \vec{w} \cdot \phi(x) = (0, 0, 1) \cdot (x_1, x_2, x_3)$$

$$= x_3 = 0$$

if $x = y = 1$ and $x = y = -1$
 $\Rightarrow x \cdot y = 1 \cdot 1 > 0$ or $-1 \cdot -1 > 0$

if $x = 1, y = -1$ or $x = -1, y = 1$
 \Rightarrow class = 1

$\Rightarrow x \cdot y = -1 \cdot 1 < 0$ or $1 \cdot -1 < 0$

hence XOR by SVM \Rightarrow class = -1

$$k(x_i, x_j) = (1 + x^T \cdot x')^2$$

$$\text{where } x^T = [x_1, x_2]$$

$$\text{and } x' = \begin{bmatrix} x'_1 \\ x'_2 \end{bmatrix}$$

$$\Rightarrow x^T \cdot x' = [x_1, x_2] \begin{bmatrix} x'_1 \\ x'_2 \end{bmatrix} = x_1 x'_1 + x_2 x'_2$$

$$\Rightarrow (1 + x_1 x'_1 + x_2 x'_2)^2$$

$$= (1 \cdot 1) + (x_1^2 x'^2_1) + (x_2^2 x'^2_2) + 2(x_1 x'_1) + 2(x_2 x'_2) + 2(x_1 x'_1)(x_2 x'_2)$$

$$\phi(x) \cdot \phi(x') = k(x_i, x_j)$$

As we have 6 terms in the expansion of kernel function, each term can be expressed as the product of 2 terms.

$$1 = 1 \cdot 1, \quad x_1^2 x'^2_1 = x_1 \cdot x_1 \cdot x'_1 \cdot x'_1$$

$$(x_2 \cdot x_2) \cdot (x'_2 \cdot x'_2) = x_2 \cdot x_2 \cdot x'_2 \cdot x'_2$$

$$2x_1 x'_1 = \sqrt{2} x_1 \cdot \sqrt{2} x'_1$$

$$2x_1x_2' = \sqrt{2}x_1 \times \sqrt{2}x_2'$$

$$2x_1x_2' = \sqrt{2}x_1x_2 \times \sqrt{2}x_1'x_2'$$

where each term is a product of one term in x & one term in x' .

$$\Rightarrow k(x, x') = [1, x_1^2, x_2^2, \sqrt{2}x_1, \sqrt{2}x_2, \sqrt{2}x_1x_2]$$

$$\cdot [1, x_1'^2, x_2'^2, \sqrt{2}x_1', \sqrt{2}x_2', \sqrt{2}x_1'x_2']$$

$$\Rightarrow \phi(x) = [1, x_1^2, x_2^2, \sqrt{2}x_1, \sqrt{2}x_2, \sqrt{2}x_1x_2]$$

$$\Rightarrow \phi(x') = [1, x_1'^2, x_2'^2, \sqrt{2}x_1', \sqrt{2}x_2', \sqrt{2}x_1'x_2']$$

There are 6 features in $\phi(x)$

$$Q5) w = \sum_{i=1}^n x_i y_i x_i = \sum_{i=1}^{10} x_i y_i x_i$$

$$= 0.414 \times 1 \times (4, 2, -1) + 0 + 0 + (0.18 \times -1 \times (2, 5, 1))$$

$$+ 0 + 0 + (0.018 \times 1 \times (3, 5, 4)) + 0$$

$$+ (0.414 \times -1 \times (2, 2, -1)) + 0$$

$$= \begin{bmatrix} 1.656 \\ 1.2006 \end{bmatrix} + \begin{bmatrix} -2.95 \\ -1.18 \end{bmatrix} + \begin{bmatrix} 0.063 \\ 0.072 \end{bmatrix}$$

$$+ \begin{bmatrix} -0.828 \\ -0.8694 \end{bmatrix} = \begin{bmatrix} -2.059 \\ -0.7768 \end{bmatrix}$$

2) Support vectors $\Rightarrow x_i \neq 0$

$\Rightarrow x_i = (4, 2, -1), (2, 5, 1), (3, 5, 4), (2, 2, -1)$ - all support vectors.

$$Q5) \quad W = \sum_{i=1}^n \lambda_i y_i x_i = \sum_{i=1}^{10} \lambda_i y_i x_i$$

$$= 0.4144 \times (4, 2.9) + 0 + 0 \\ + (1.18 \times -1 \times (2.5, 1)) + 0 + 0 \\ + (0.18 \times 1 \times (3.5, 4)) + 0 + \\ (0.4144 \times -1 \times (2, 2.1)) + 0$$

$$= \begin{bmatrix} 1.656 \\ 1.2006 \end{bmatrix} + \begin{bmatrix} -2.95 \\ -1.18 \end{bmatrix} + \begin{bmatrix} 4.13 \\ 4.72 \end{bmatrix} + \begin{bmatrix} -0.828 \\ -0.864 \end{bmatrix}$$

$$W = \begin{bmatrix} 2.008 \\ 3.876 \end{bmatrix}$$

Support vectors $\Rightarrow \lambda_i \neq 0$
 $\Rightarrow x_i = (4, 2.9), (2.5, 1), (3.5, 4), (2, 2.1)$ are all support vectors.

Using KKT cond

$$b_i = 1 - W^T x_i^{(S)} \times y_i$$

$$b_1 = 1 - W^T \begin{bmatrix} 4 \\ 2.9 \end{bmatrix} = -18.2724$$

$$b_2 = -1 - W^T \begin{bmatrix} 2.5 \\ 1 \end{bmatrix} = 9.896$$

$$b_3 = 1 - W^T \begin{bmatrix} 3.5 \\ 4 \end{bmatrix} = -21.508$$

$$b_4 = -1 - W^T \begin{bmatrix} 2 \\ 2.1 \end{bmatrix} = -13.1556$$

$$\Rightarrow Avg = -15.708$$

for (3,3)

$$w^T x + b = 17.652 - 15.708$$

$$= 1.944 > 0$$

\Rightarrow belongs to class
1

Q6) Bonus

$$\text{RBF kernel} = e^{-\gamma \|U-V\|^2}$$

$$\|U-V\|^2 = \sqrt{(u_1-v_1)^2 + (u_2-v_2)^2 + \dots}$$

but U & V are 1-D

$$\Rightarrow \|U-V\| = \sqrt{(U-V)^2} = (U-V)$$

$$\Rightarrow \text{kernel} = e^{-\gamma (U-V)^2} = \phi\left(\frac{U}{\sqrt{\gamma}}\right) \phi\left(\frac{V}{\sqrt{\gamma}}\right)$$

$$= e^{-\gamma (U^2 + V^2 - 2UV)}$$

$$= e^{-\gamma (-U^2 - V^2 + 2UV)}$$

$$= e^{\gamma (-U^2 - V^2)} \times e^{\gamma (2UV)}$$

using Taylor expansion of $e^{\gamma (2UV)}$

$$= \left(1 + \frac{\gamma 2UV}{1!} + \frac{(\gamma 2UV)^2}{2!} + \dots \right)$$

$$= 1 + \frac{\sqrt{2\gamma}U \cdot \sqrt{2\gamma}V}{1!} + \frac{(\sqrt{2\gamma})^2 U^2 \times \sqrt{2\gamma} \cdot V^2}{\sqrt{2!} \sqrt{2!}} + \dots$$

where each term is a dot product of terms from U & from V .

$$\Rightarrow k(U, V) = e^{-U^2\gamma} \times e^{-V^2\gamma} \times \left(1 + \frac{\sqrt{2\gamma}U \cdot \sqrt{2\gamma}V}{1!} + \dots \right)$$

$$= e^{-U^2\gamma} \left(1, \frac{\sqrt{2\gamma}U}{1!}, \frac{(\sqrt{2\gamma})^2 U^2}{\sqrt{2!}} \dots \right) \cdot \left(e^{-V^2\gamma} \left(1, \frac{\sqrt{2\gamma}V}{1!}, \dots \right) \right)$$

$$\Rightarrow \phi(u) = e^{-vu^2} \left(1, \frac{\sqrt{2v}u}{1!}, \frac{\sqrt{(2v)^2}u^2}{2!}, \dots \right)$$

$$\phi(v) = e^{-rv^2} \left(1, \frac{\sqrt{2r}v}{1!}, \frac{\sqrt{(2r)^2}v^2}{2!}, \dots \right)$$

$$\Rightarrow K(u, v) = \phi(u) \cdot \phi(v)$$

$$\Rightarrow \text{W-C transform}$$

$$\phi(x) = e^{-rx^2} \left(1, \frac{\sqrt{2r}x}{1!}, \frac{\sqrt{(2r)^2}x^2}{2!}, \dots \right)$$

as there is a factorial in the denominator,
when the ~~the~~ degree of the terms ≈ 1000000 ,
then the denominator is going to be $\approx \infty$,
but the numerator is although quite big, it is not
as big as the denominator. \Rightarrow the terms go to
zero.

Hence higher degrees are close to zero (0) in
2bf kernel.